

最佳跨架構AI應用程式平台

Joel Lin, Intel
Dec. 28th, 2021

Choose the Best Accelerated Technology

Intel[®] oneAPI Tools for AI

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Dec. 28th, 2021



Intel's oneAPI Ecosystem

Built on Intel's Rich Heritage of CPU Tools Expanded to XPU

oneAPI

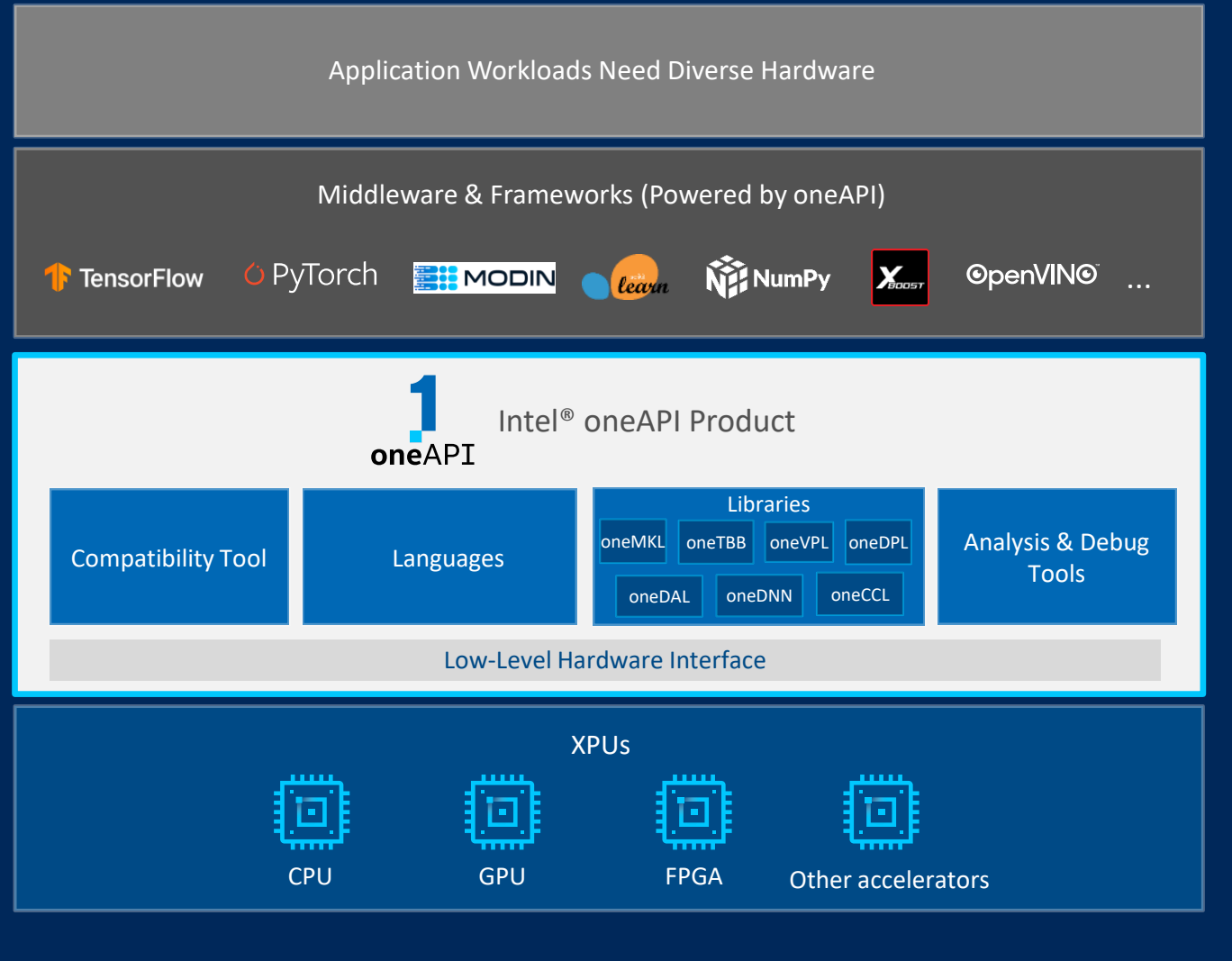
A cross-architecture language based on C++ and SYCL standards

Powerful libraries designed for acceleration of domain-specific functions

A complete set of advanced compilers, libraries, and porting, analysis and debugger tools

Powered by oneAPI

Frameworks and middleware that are built using one or more of the oneAPI industry specification elements, the DPC++ language, and libraries listed on oneapi.com.



[Available Now](#)

Intel oneAPI Software Tools for AI & Analytics

Intel® oneAPI Toolkits 2022 is released this December.
Find out more details on the [online release notes](#).

Intel® oneAPI Toolkits



Intel® oneAPI AI Analytics Toolkit

Accelerate machine learning & data science pipelines with optimized deep learning frameworks & high-performing Python libraries

[Data Scientists, AI Researchers, DL/ML Developers](#)



Intel® oneAPI Base Toolkit

Incl. Intel® oneAPI Deep Neural Network Library (oneDNN), Intel® oneAPI Collective Communications Library (oneCCL), & Intel® oneAPI Data Analytics Library (oneDAL)

Optimize primitives for algorithms and framework development

[DL Framework Developers - Optimize algorithms for Machine Learning & Analytics](#)

Toolkit Powered by oneAPI

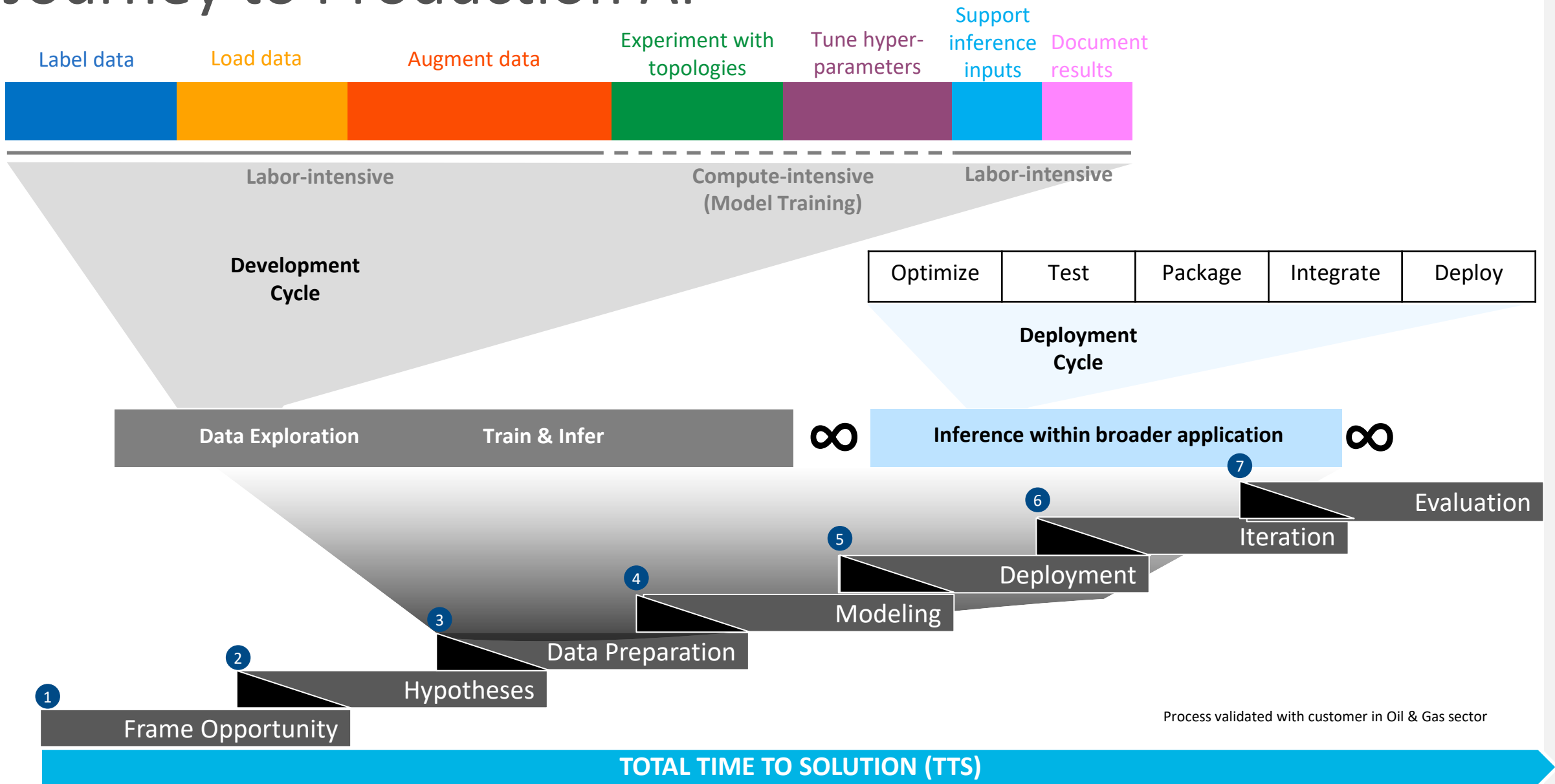
Intel® Distribution of OpenVINO™ Toolkit

Deploy high performance inference & applications from edge to cloud

[AI Application, Media, & Vision Developers](#)

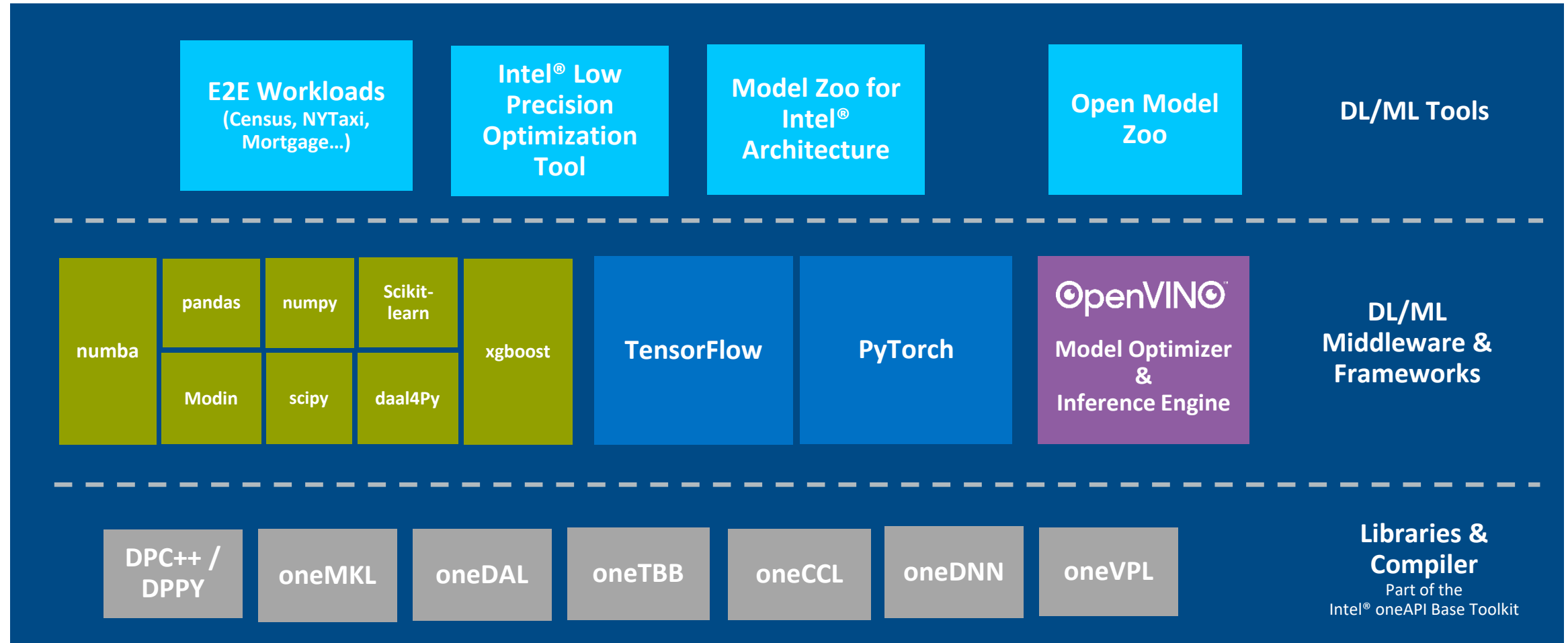


Journey to Production AI



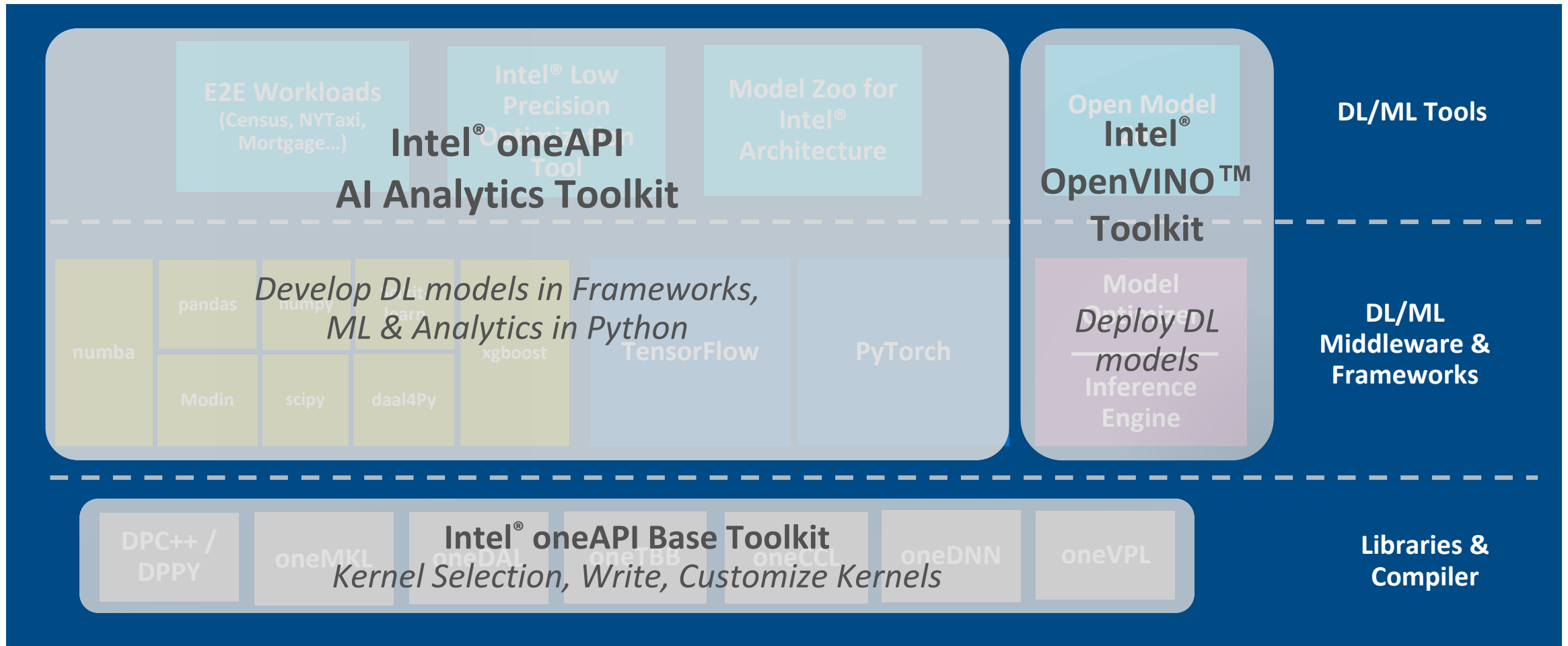
AI Software Stack for Intel XPU

Intel offers a Robust Software Stack to Maximize Performance of Diverse Workloads



AI Software Stack for Intel XPU

Intel offers a Robust Software Stack to Maximize Performance of Diverse Workloads



Full Set of Intel oneAPI cross-architecture AI ML & DL Software Solutions

Intel® oneAPI Base Toolkit

Accelerate Data-centric Workloads

A set of core tools and libraries for developing high-performance applications on Intel® CPUs, GPUs, and FPGAs

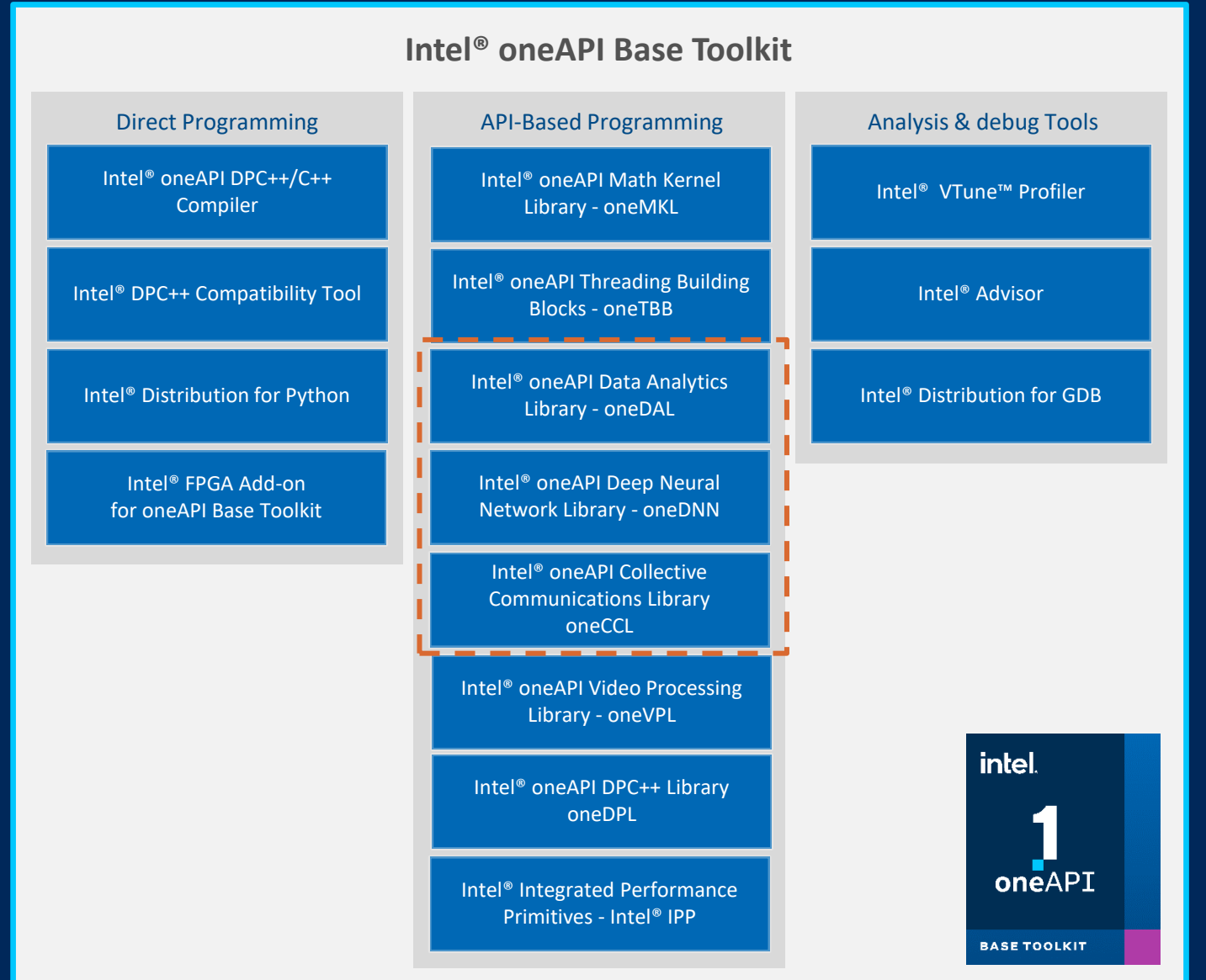
Intel® oneAPI IoT Toolkit 2022 is out. Find out more details on the release notes.

Who Uses It?

- A broad range of developers across industries
- Native Code Developers/Framework Developers

Top Features/Benefits

- Data Parallel C++ (DPC++) compiler, library and analysis tools; DPC++ Compatibility tool helps migrate existing code written in CUDA
- Optimized performance libraries for threading, math, data analytics, deep learning, and video/image/signal processing



AI Development Cycle

Intel® oneAPI AI Analytics Toolkit

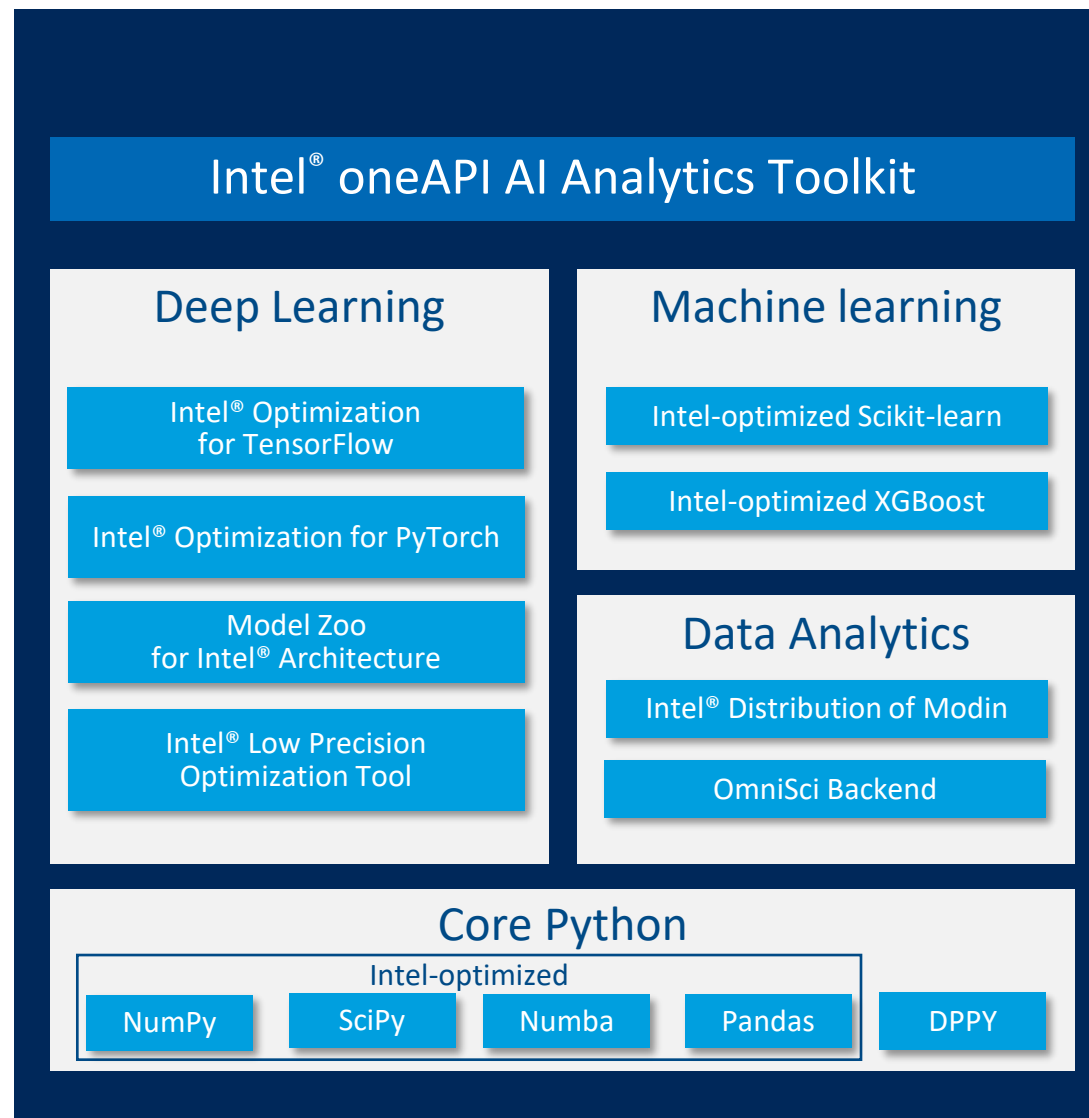
Accelerates end-to-end Machine Learning and Data Analytics pipelines with frameworks and libraries optimized for Intel® architectures

Target Model building, evaluating, training and datap preprocessing and data analytics in AI Development Cycle

Who Uses It?

Data scientists, AI Researchers, Machine and Deep Learning developers, AI application developers

Learn More: intel.com/oneAPI-AIKit



Performance Benefits

Maximize Hardware Value with Intel-optimized Software

Deep Learning Training & Inference Performance

Uses Intel® Optimization for PyTorch with 3rd Gen Intel® Xeon® Scalable Processors

Training	# Cores per instance	# Instances	BF16 (samples/s)	FP32 (samples/s)	Speedup Ratio
DLRM	28	1	99321	71061	1.40
ResNet-50	28	4	399	243	1.64
ResNeXt-101 32x4d	28	4	193	120	1.60

Table 1. BF16 training performance gains over baseline (FP32 with Intel oneDNN)

Inference	# Cores per instance	# Instances	INT8 (samples/s)	FP32 (samples/s)	Speedup Ratio
DLRM	1	28	611082	214559	2.85

Table 2. INT8 inference performance gains over baseline (FP32 with Intel oneDNN)

ML Performance with Intel-optimized scikit-learn

```
from sklearn.svm import SVC
X, Y = get_dataset()

clf = SVC().fit(X, y)
res = clf.predict(X)
```

Common Scikit-learn (mainline)

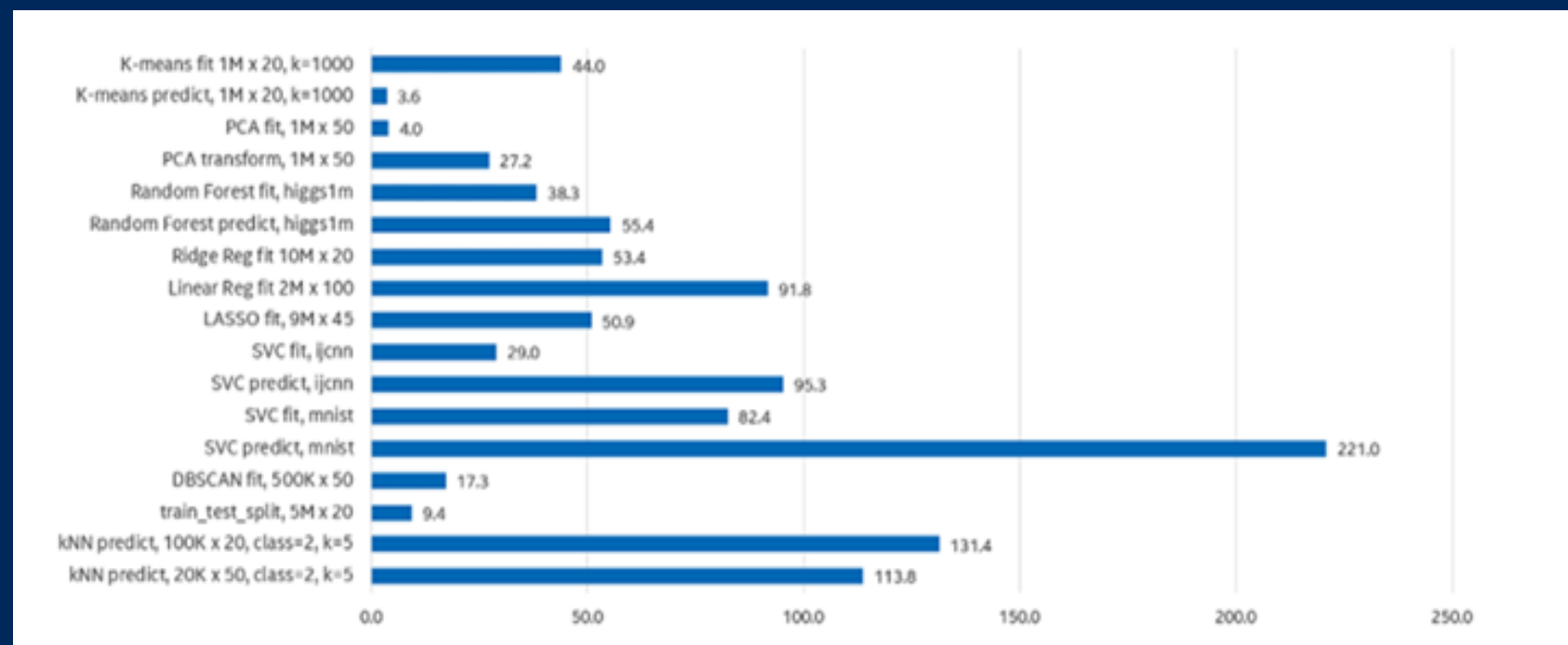
```
import daal4py as d4p
d4p.patch_sklearn()

from sklearn.svm import SVC
X, Y = get_dataset()

clf = SVC().fit(X, y)
res = clf.predict(X)
```

Scikit-learn on Intel CPU optimized
by Intel® oneAPI AI Analytics Toolkit

Stock scikit-learn vs Intel-optimized scikit-learn



Same Code, Same Behavior

Scikit-learn, not scikit-learn-like

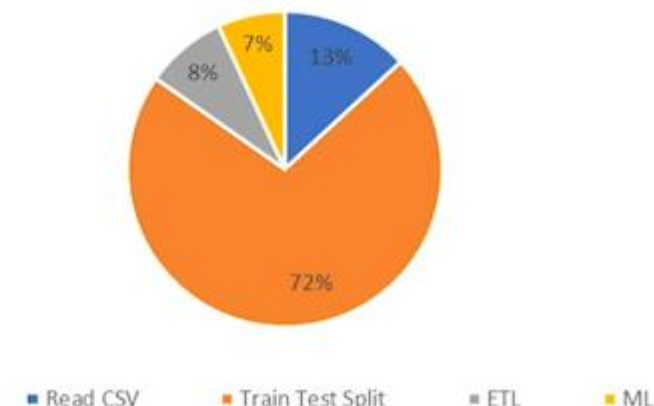
End-to-End Data Pipeline Acceleration

- **Workload:** Train a model using 50 years of Census dataset from IPUMS.org to predict income based on education
- **Solution:** Intel Modin for data ingestion and ETL, Daal4Py and Intel scikit-learn for model training and prediction
- **Performance Gains**
 - Read_CSV (Read from disk and store as a dataframe): **6x**
 - ETL operations: **38x**
 - Train Test Split: **4x**
 - ML training (fit & predict) with Ridge Regression: **21x**

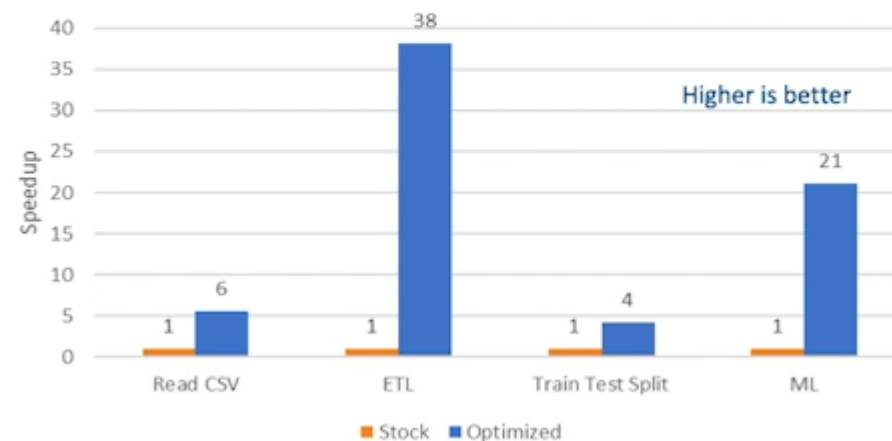
End-to-End Census Workload Performance

Tested by Intel as of 10/15/2020. 2x Intel® Xeon® Platinum 8280 @ 28cores, OS: Ubuntu 19.10.5.3.0-64-generic Mitigated, 384GB RAM. SW: Modin 0.8.1, scikit-learn 0.22.2, Pandas 1.0.1, Python 3.8.5, Daal4Py 2020.2 Census Data, (21721922, 45). Dataset is from IPUMS USA, University of Minnesota, www.ipums.org . Version 10.0.

End-to-End Time Breakdown : Census Education to Income



End-to-End Census: Speedup with optimized libraries



AI Deployment Cycle

High-Performance Deep Learning Using Intel® Distribution of OpenVINO™ toolkit - Powered by oneAPI

A toolkit for fast, more accurate real-world results using high-performance AI and computer vision inference deployed into production on Intel XPU architectures (CPU, GPU, FPGA, VPU) from edge to cloud

Who needs this product?

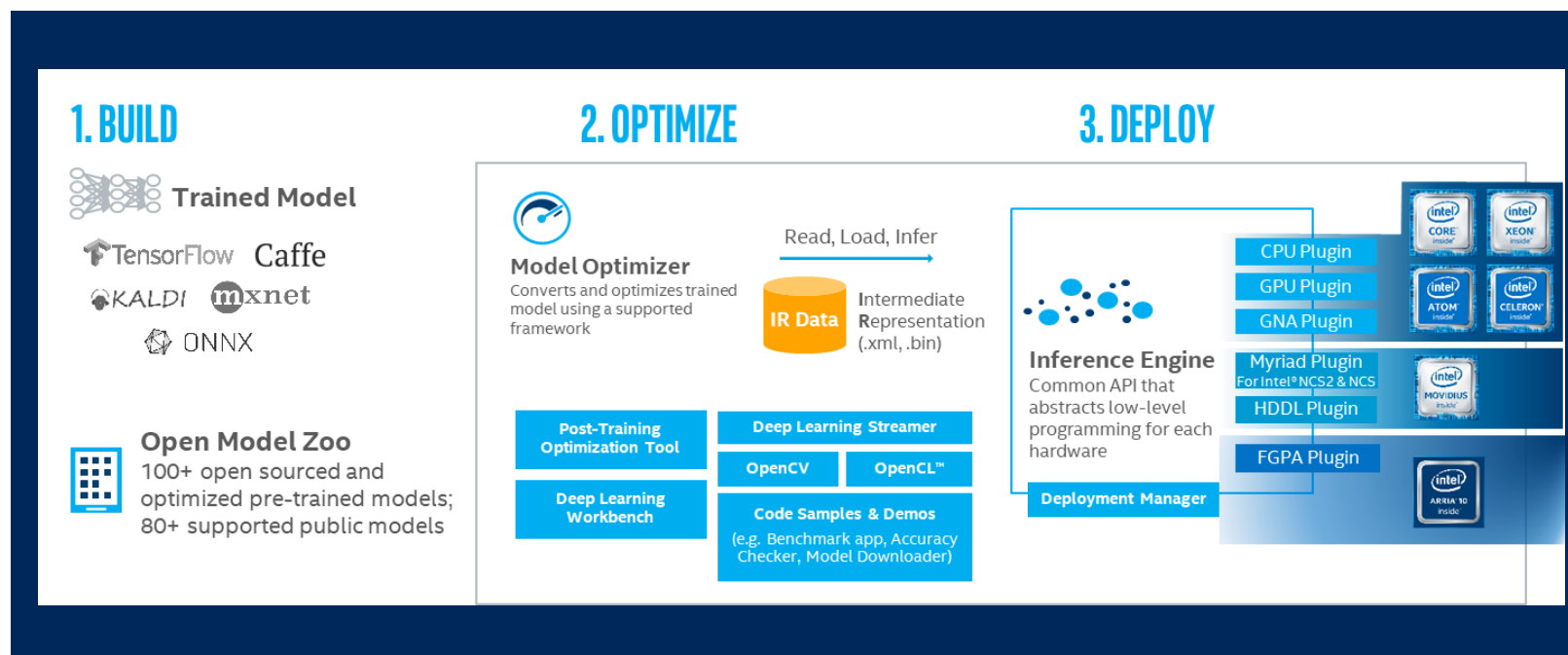
AI application developers, OEMs, ISVs, System Integrators, Vision and Media developers

Top Features/Benefits

High-performance, deep learning inference deployment

Streamlined development; ease of use

Write once, deploy anywhere



Proven, industry-leading accelerated technology

software.intel.com/openvino-toolkit

Which Toolkit Should I Use

Use Both!

Intel® oneAPI Analytics Toolkit & Intel® Distribution of OpenVINO™ toolkit

Toolkits are complementary to each other and recommendation is to use them both based on your current phase of AI Journey

- I am **exploring and analyzing data**; I am **developing models**
- I want **performance and compatibility** with frameworks and libraries I use
- I would like to have **drop-in acceleration** with little to no additional code changes
- I prefer **not to learn any new tools or languages**



Data Scientist/ML Developer
Intel® oneAPI AI Analytics Toolkit

- I am **deploying models**
- I want **leading performance and efficiency** across multiple target HW
- I'm concerned about **having lower memory footprint**, which is critical for deployment
- I am **comfortable with learning and adopting a new tool or API** to do so



App Developer
Intel® Distribution of OpenVINO™ toolkit

If you prefer working on primitives and to optimize kernels and algorithms directly using oneAPI libraries (oneDNN, oneCCL & oneDAL), then use [Intel® oneAPI Base Toolkit](#)

Accrad AI-based Solution Helps Accelerate Lung Disease Diagnosis

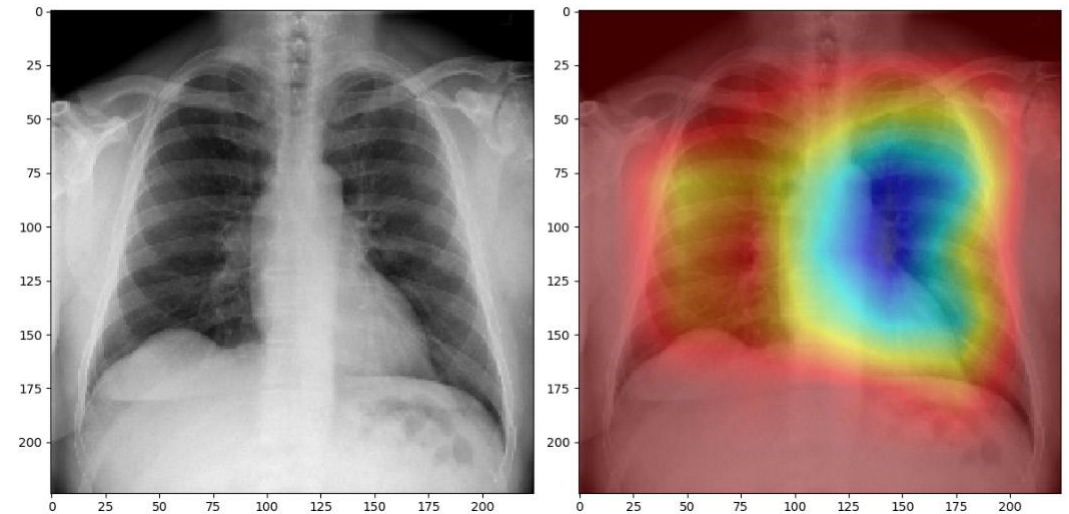
Optimized by Intel® oneAPI Analytics Toolkit & Intel® Distribution of OpenVINO™ toolkit

AI Machine Vision Disease Detection

CheXRad is a machine learning edge application that helps radiologists and physicians identify COVID-19, viral pneumonia and other diseases on chest X-ray images, and predict the need for ventilators.

- *CheXRad* comes pre-configured with a COVID-19 and viral pneumonia classification neural network.
- To architect, train and validate the neural network, Accrad used Intel Tensorflow from AI Analytics Toolkit and the infrastructure provided by Intel oneAPI DevCloud to develop the model.
- To optimize its model for deployment, Accrad used OpenVINO™ toolkit and Intel® DevCloud for Edge.
- *CheXRad* could label pathologies in 140 chest x-rays in just **90 seconds**—up to **160x faster** than radiologists, at comparable levels of accuracy, sensitivity and specificity.

Ground Truth Class: 0 (non-COVID-19)
Predicted Class: 0 (non-COVID-19)
Prediction probabilities: ['1.00', '0.00']



*“With the help of Intel, we were able to **train, optimize, and deploy** a machine learning model in **less time and at a lower operational cost** than available alternatives, enabling us to get to market fast with a powerful solution that’s optimized for Intel® architecture.” – Moloti Nakampe, R&D Director*

Learn more in this [solution brief](#)

AbbVie Machine Translation Solution

Optimized by Intel® oneAPI Analytics Toolkit & Intel® Distribution of OpenVINO™ toolkit

A research-based biopharmaceutical company powered by Intel® Xeon® processors.

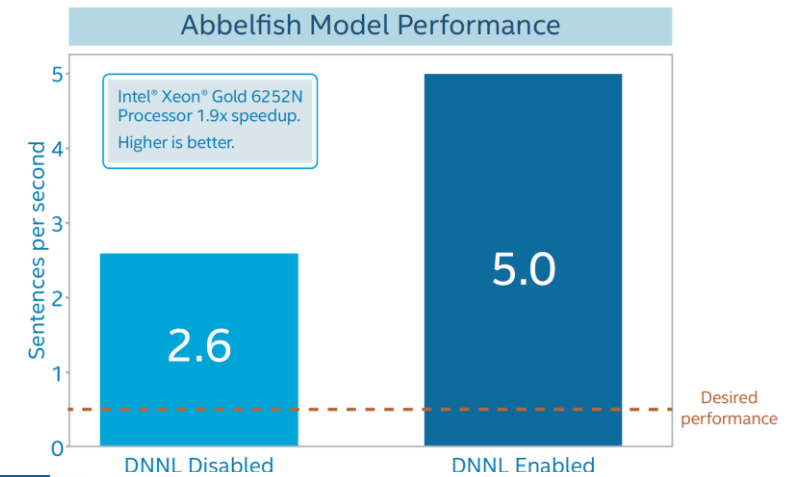
Abbelfish Machine Translation uses Intel® Optimization for TensorFlow of AI Analytics Toolkit

- A custom model that provides more accurate translations than commercially available ones. Model includes 24 layers and 500+ million parameters that took more than 4 months to train
- Intel TensorFlow provided a greater performance boost while AbbVie did not have to change its code or APIs from standard TensorFlow

AbbVie Search uses Intel® Distribution of OpenVINO™ toolkit

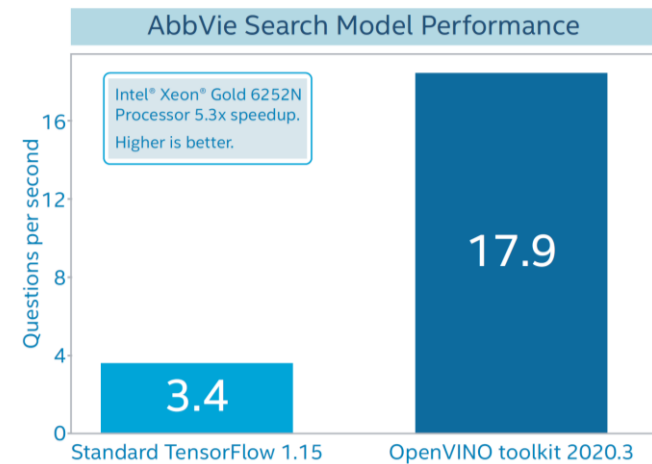
- OpenVINO toolkit provided great speed-up to answer questions from a scientific article or clinical report when compared to standard TensorFlow
- Requires scaling across the company, so uses OpenVINO™ Model Server to serve inferences

Learn more: [Intel News Byte Dec. 15, 2020](#)



Drop-in acceleration

Figure 3. AbbVie's Abbelfish translated over five sentences per second using Intel Optimization for TensorFlow with oneAPI Deep Neural Network Library (oneDNN).¹



Deployment

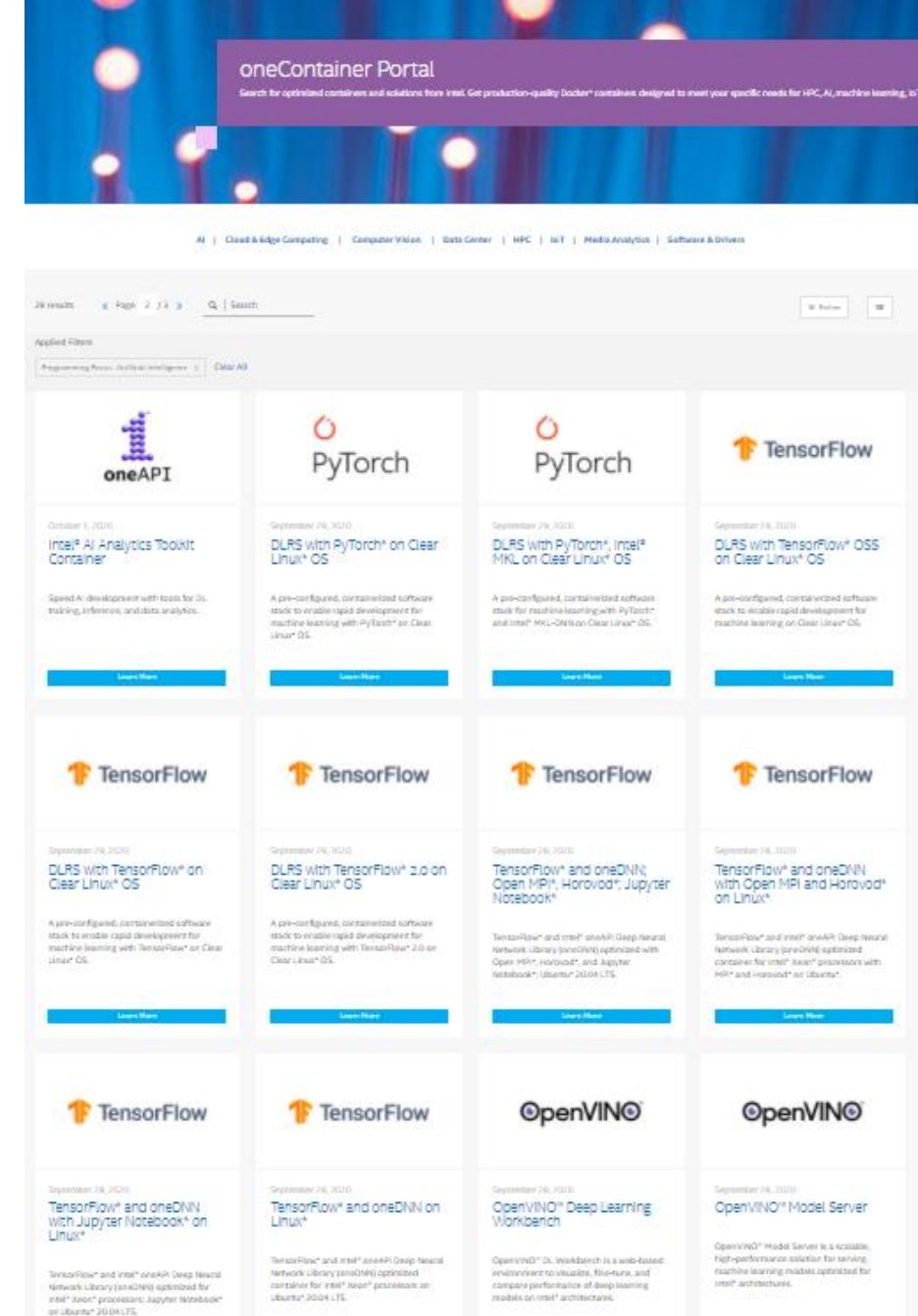
Figure 5. Comparison of AbbVie Search inference between unoptimized TensorFlow 1.15 (oneDNN disabled) and OpenVINO toolkit 2020.3.¹

AI Containers for Flexibility

- Optimized, validated, deployable AI containers and artifacts for Intel® platforms
- Available via Docker containers and Intel software stacks. Will expand to include Kubernetes orchestrations, Helm charts, AI models, pipelines and more
- [Access oneContainer Portal](#)

Key Models (GPU-WIP, Check the portal for latest availability)

Topology	Frameworks	Topology	Framework
DLRM	PYT	Mask R-CNN	PYT, TF, OV
ResNet50	PYT, TF, OV	RNN-T	PYT, TF, OV
BERT-large	PYT, TF, OV	3D-UNet	TF, OV
Transformer-LT	PYT, TF	DIEN	TF
MobileNet-v1	PYT, TF, OV	Wide & Deep	PYT, TF
SSD-Mobilenet-v1	PYT, TF, OV	RNX101	
SSD-Resnet34	PYT, TF, OV	Yolo-V3	PYT, TF, OV
WaveNet*	TF	NCF*	TF



Key Takeaways & Call to Action

- Intel toolkits are FREE, complementary & work seamlessly together
- They help achieve performance & efficiency across different stages of AI Journey
- Recommend the toolkits based on current phase of customer pipeline

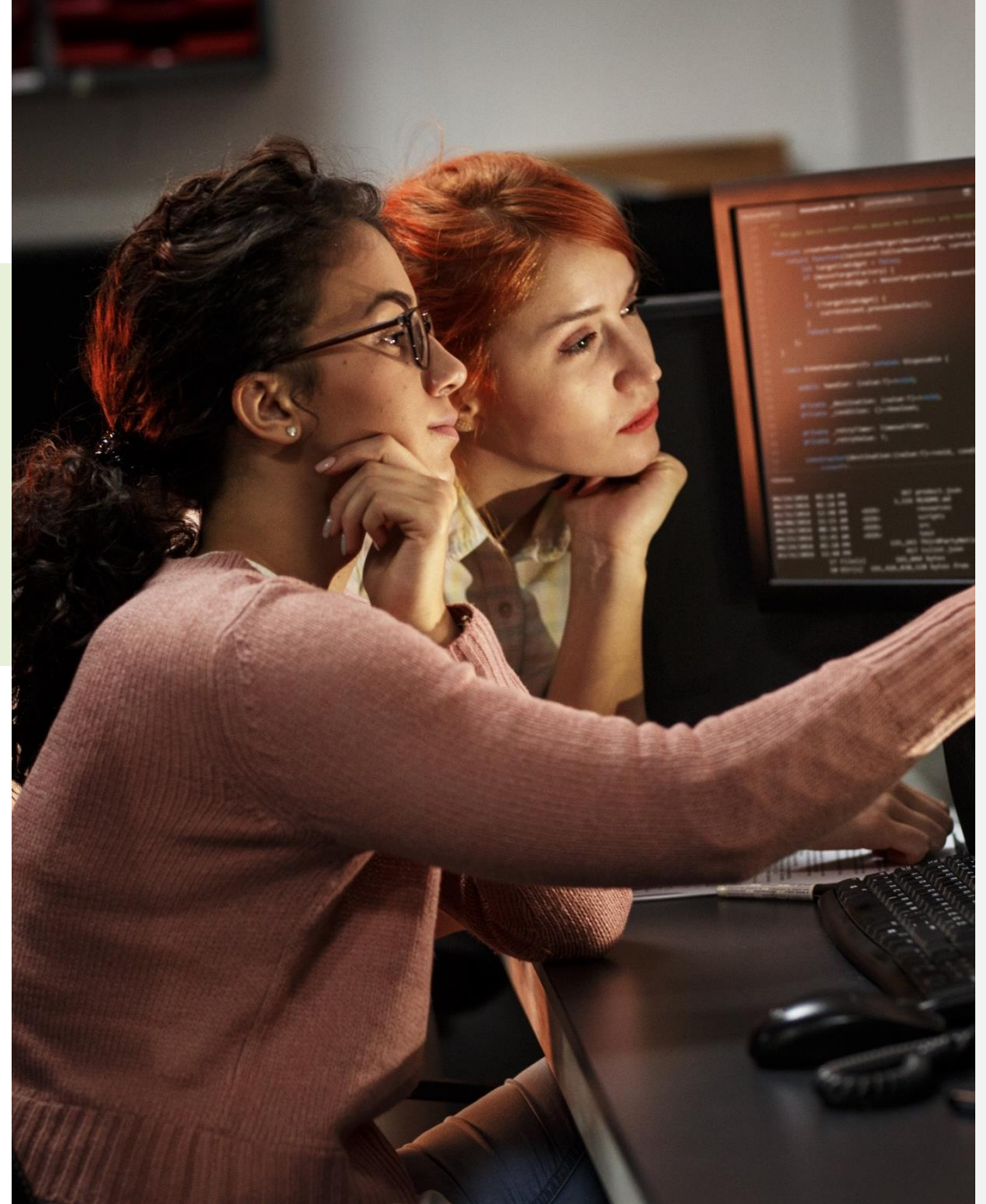
Download the toolkits

[Intel® oneAPI AI Analytics Toolkit](#)

[Intel® Distribution of OpenVINO™ toolkit](#)

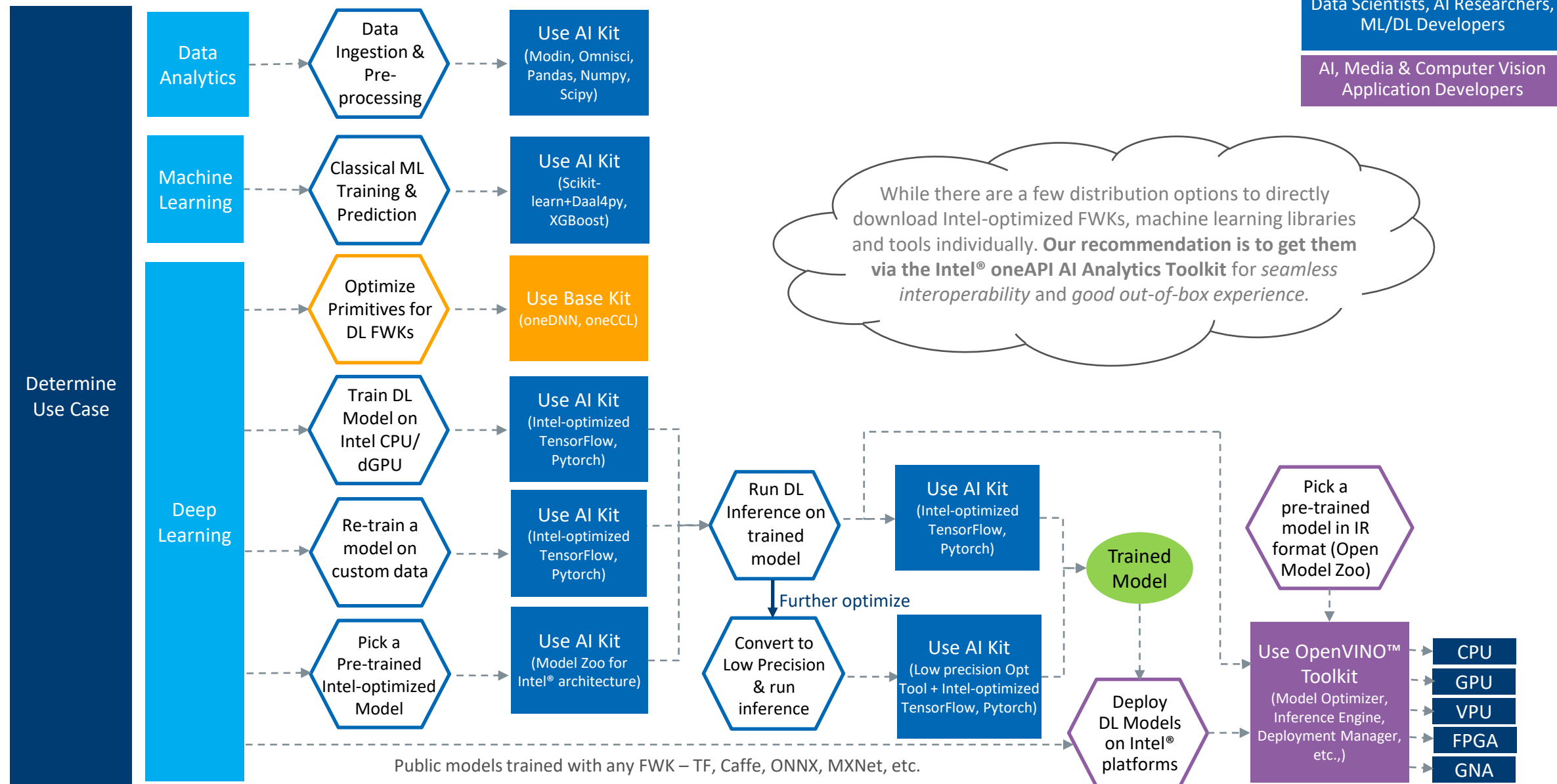
[Intel® oneAPI Base Toolkit](#)

Learn more about [Intel® oneAPI Toolkits](#)
intel.com/oneAPI-AIToolkits

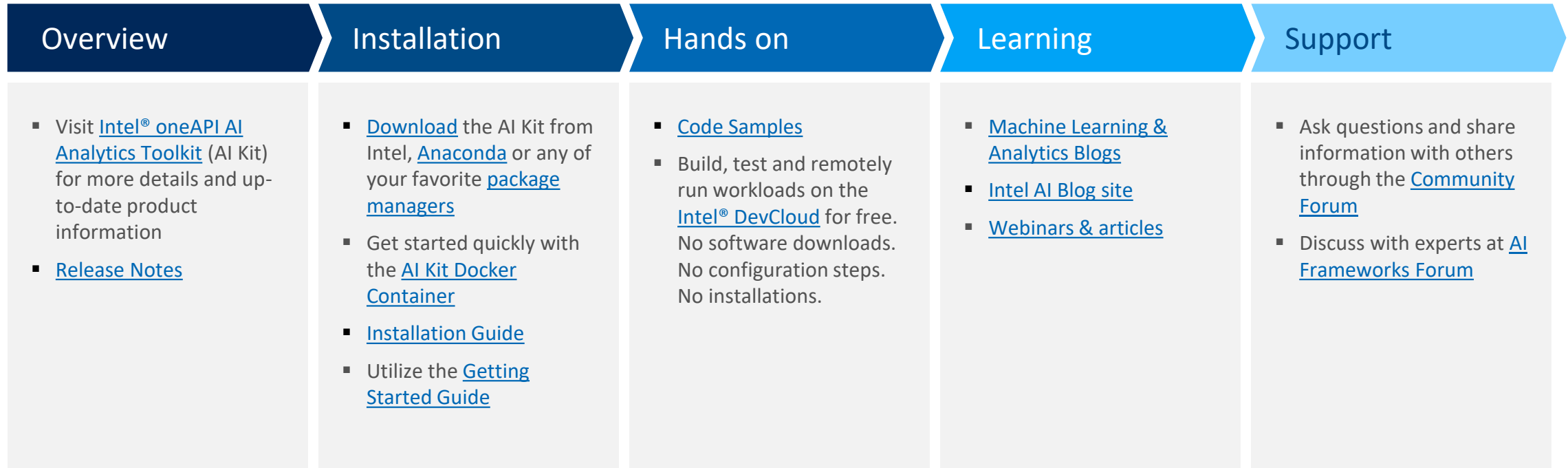


BackUp

AI Development Workflow

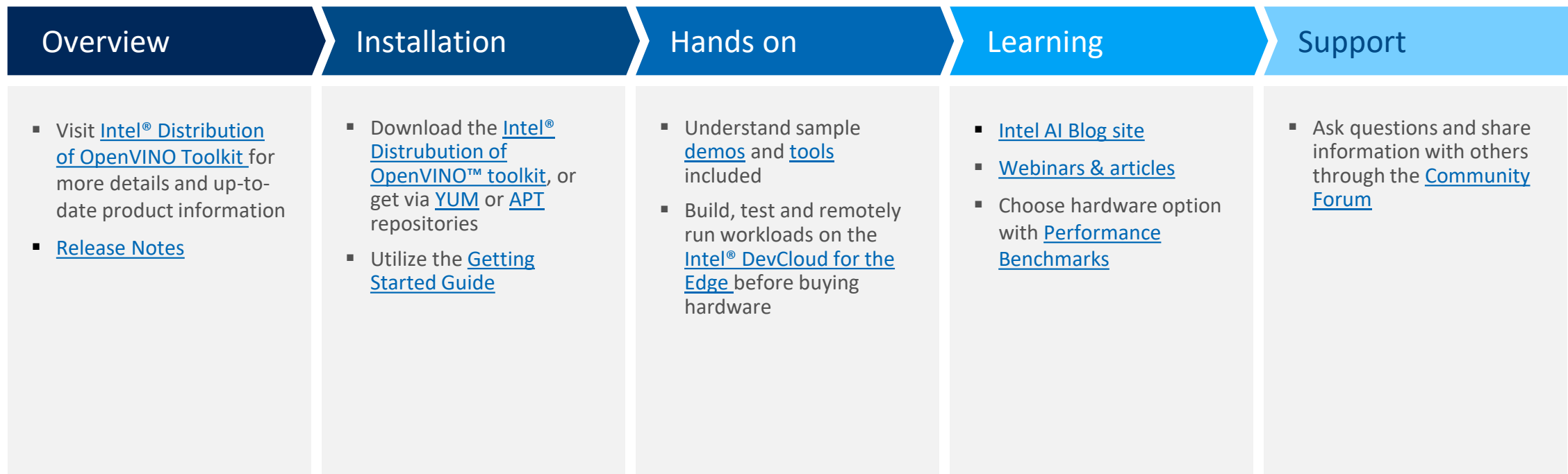


Getting Started with Intel® oneAPI AI Analytics Toolkit



Download Now

Getting Started with Intel® Distribution of OpenVINO™ Toolkit



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Performance results are based on testing as of dates shown in configurations and may not reflect all publicly available updates. See backup for configuration details. No product or component can be absolutely secure.

Your costs and results may vary.

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Intel does not control or audit third-party data. You should consult other sources to evaluate accuracy.

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Configurations

Deep Learning Training and Inference Performance using Intel® Optimization for PyTorch with 3rd Gen Intel® Xeon® Scalable Processors

ResNet50/ResNext101 (FP32/BF16): batch size 128/instance, 4 instances.

ResNet50/ResNext101 dataset (FP32/BF16): [ImageNet Dataset](#)

DLRM batch size (FP32/BF16): 2K/instance, 1 instance

DLRM dataset (FP32/BF16): [Criteo Terabyte Dataset](#)

DLRM batch size (INT8): 16/instance, 28 instances, dummy data.

Tested by Intel as of 6/2/2020.

Intel® Xeon® Platinum 8380H Processor, 4 socket, 28 cores HT On Turbo ON Total Memory 768 GB (24 slots/ 32GB/ 3200 MHz), BIOS: WLYDCRB1.SYS.0015.P96.2005070242 (ucode: 0x700001b),

Ubuntu 20.04 LTS, kernel 5.4.0-29-generic

PyTorch: <https://github.com/pytorch/pytorch.git>

Intel Extension for PyTorch: <https://github.com/intel/intel-extension-for-pytorch.git>

gcc: 8.4.0,

Intel® oneAPI Deep Neural Network Library (oneDNN) version: v1.4

ResNet50: <https://github.com/intel/optimized-models/tree/master/pytorch/ResNet50>

ResNext101 32x4d: https://github.com/intel/optimized-models/tree/master/pytorch/ResNext101_32x4d

DLRM: <https://github.com/intel/optimized-models/tree/master/pytorch/dlrm>

Inference Throughput FP32 vs Int8 optimized by Intel® Optimization for Tensorflow and Intel® Low Precision Optimization Tool (part of the Intel® oneAPI AI Analytics Toolkit)

Tested by Intel as of : 10/26/2020: TensorFlow v2.2 (<https://github.com/Intel-tensorflow/tensorflow/tree/v2.2.0>); Compiler: GCC 7.2.1; DNNL(<https://github.com/oneapi-src/oneDNN>) v1.2.0

75d0b1a7f3586c212e37acebbb8acd221cee7216; Dataset: ImageNet/Coco/Dummy, refer to each model README; Precision: FP32 and Int8

Platform: Intel® Xeon® Platinum 8280 CPU; #Nodes: 1; #Sockets: 2; Cores/socket: 28; Threads/socket: 56; HT: On; Turbo: On; BIOS version: SE5C620.86B.02.01.0010.010620200716; System DDR

Mem Config: 12 slots / 16GB / 2933; OS: CentOS Linux 7.8; Kernel: 4.4.240-1.el7.elrepo.x86_64

Stock scikit-learn vs Intel-optimized scikit-learn

Testing by Intel as of 10/23/2020. Intel® oneAPI Data Analytics Library 2021.1 (oneDAL), scikit-learn 0.23.1, Intel® Distribution for Python 3.8; Intel® Xeon® Platinum 8280LCPU @ 2.70GHz,

2Sockets, 28 cores per socket, 10M samples, 10 features, 100 clusters, 100 iterations, float32

XGBoost CPU vs GPU

Test configs: Tested by Intel as of 10/13/2020;
CPU: c5.18xlarge AWS Instance (2 x Intel® Xeon Platinum 8124M @ 18 cores, OS: Ubuntu 20.04.2 LTS, 193 GB RAM. GPU: p3.2xlarge AWS Instance (GPU: NVIDIA Tesla V100 16GB, 8 vCPUs), OS: Ubuntu 18.04.2 LTS, 61 GB RAM. SW: XGBoost 1.1:build from sources. compiler – G++ 7.4, nvcc 9.1. Intel® Data Analytics Acceleration Library (Intel® DAAL): 2019.4 version; Python env: Python 3.6, Numpy 1.16.4, Pandas 0.25, Scikit-learn 0.21.2.

XGBoost fit CPU acceleration

Test configs: Tested by Intel as of 10/13/2020; c5.24xlarge AWS Instance, CLX 8275 @ 3.0GHz, 2 sockets, 24 cores per socket, HT:on, DRAM (12 slots / 32GB / 2933 MHz); SW: XGBoost 0.81, 0.9, 1.0 and 1.1:build from sources. compiler – G++ 7.4, nvcc 9.1. Intel® DAAL: 2019.4 version; Python env: Python 3.6, Numpy 1.16.4, Pandas 0.25, Scikit-learn 0.21.2.

End-to-End Census Workload Performance

Tested by Intel as of 10/15/2020. 2x Intel® Xeon® Platinum 8280 @ 28cores, OS: Ubuntu 19.10.5.3.0-64-generic Mitigated, 384GB RAM. SW: Modin 0.8.1, scikit-learn 0.22.2, Pandas 1.0.1, Python 3.8.5, Daal4Py 2020.2 Census Data, (21721922, 45). Dataset is from IPUMS USA, University of Minnesota, www.ipums.org . Version 10.0.

Tiger Lake + Intel® Distribution of OpenVINO™ toolkit vs Coffee Lake CPU

System Board	Intel prototype, TGL U DDR4 SODIMM RVP	ASUSTeK COMPUTER INC. / PRIME Z370-A
CPU	11 th Gen Intel® Core™ -5-1145G7E @ 2.6 GHz.	8 th Gen Intel® Core™ i5-8500T @ 3.0 GHz.
Sockets / Physical cores	1 / 4	1 / 6
HyperThreading / Turbo Setting	Enabled / On	Na / On
Memory	2 x 8198 MB 3200 MT/s DDR4	2 x 16384 MB 2667 MT/s DDR4
OS	Ubuntu* 18.04 LTS	Ubuntu* 18.04 LTS
Kernel	5.8.0-050800-generic	5.3.0-24-generic
Software	Intel® Distribution of OpenVINO™ toolkit 2021.1.075	Intel® Distribution of OpenVINO™ toolkit 2021.1.075
BIOS	Intel TGLIFUI1.R00.3243.A04.2006302148	AMI, version 2401
BIOS release date	Release Date: 06/30/2020	7/12/2019
BIOS Setting	Load default settings	Load default settings, set XMP to 2667
Test Date	9/9/2020	9/9/2020
Precision and Batch Size	CPU: INT8, GPU: FP16-INT8, batch size: 1	CPU: INT8, GPU: FP16-INT8, batch size: 1
Number of Inference Requests	4	6
Number of Execution Streams	4	6
Power (TDP Link)	<u>28 W</u>	<u>35W</u>

intel®



Thank You.